



O https://www.kaggle.com/competitions/sentiment-analysis-on-movie-reviews/leaderboard

# Sentiment Analysis on Movie Reviews

Predictive analytics with machine learning by Tiago Pedro

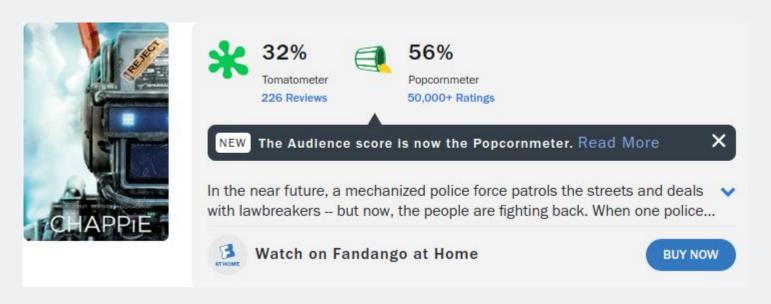


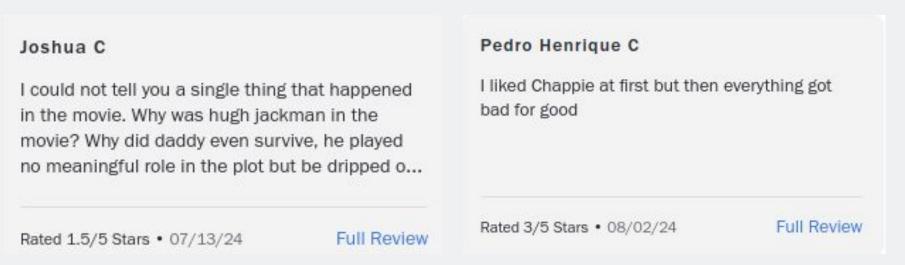




# Competition Description

For this competition we are delivered a dataset from Rotten Tomatoes, with comments and the sentiment associated with the comment, we have a training data set and a test data set for the submission.





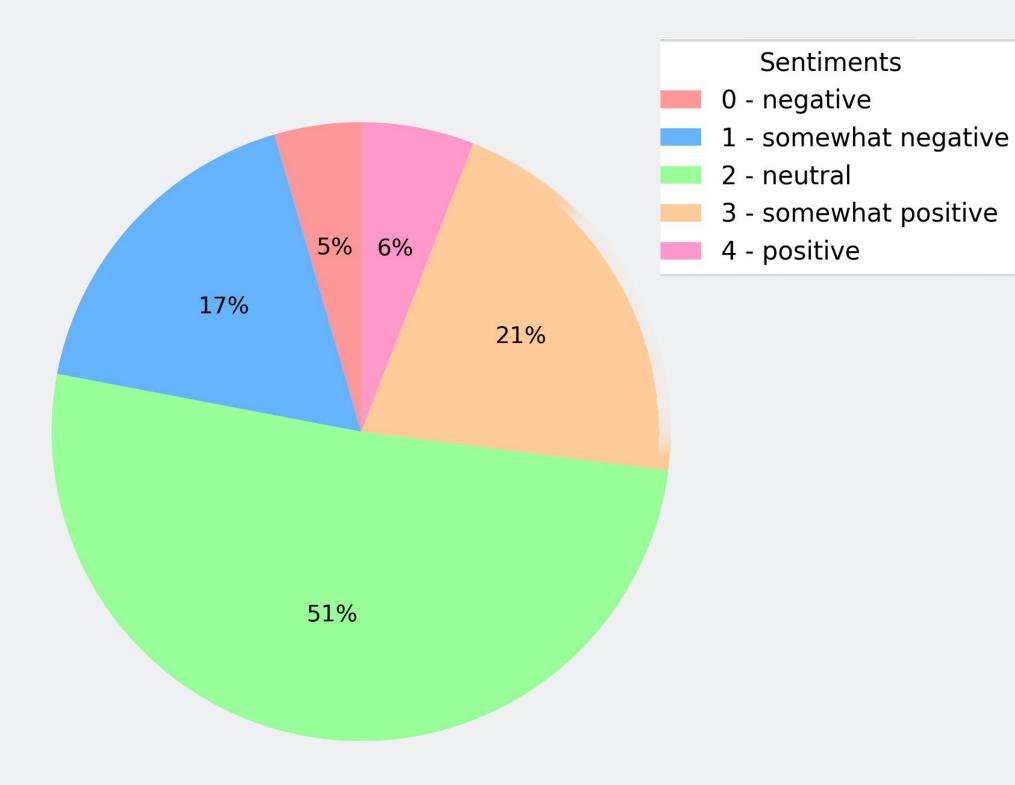






# EDA

The train dataset is composed of 4 columns, 'Phraseld', 'Sentenceld', 'Phrase' and 'Sentiment', it has 156060 rows, with no duplicated or null values, and the sentiment is a integer that ranges from 0 to 4.



Sentiments







# EDA

### 'Phraseld'

An integer with the id of the phrase, a phrase then is split in multiple sentences

### 'Sentenceld'

An integer with the number of the sentence

### 'Phrase'

The string that we will use to train the model, containing the comment or parts of it







# Data Preprocessing



### **Null values**

The data set had no Null values



### **SMOTE**

Because of the high count of Neutral sentiments, we balanced the dataframe with SMOTE



### **Spacy**

To preprocess the text, including tokenization, lemmatization, and removal of stopwords.



### **TF-IDF**

Term Frequency-Inverse Document Frequency, to convert the text into numbers







# Model Evaluation & Metrics

Model	Accuracy	MAE	RMSE	R2 score
Logistic Regression	0.57	0.49	0.80	0.66
Random Forest	N/A	0.61	0.88	0.06
SVM	0.65	0.40	0.72	0.66
KNN	0.52	0.59	0.90	0.71
XGBClassifier	0.55	0.53	0.53	0.63





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# Hyperparameter Tuning

**Logistic Regression** 

C (Inverse of Regularization Strength)

penalty (Regularization Type)

solver (Algorithm to Optimize the Model

max\_iter (Maximum Iterations)

**SVM** 

C (Regularization Parameter)

kernel (Kernel Function

gamma (Kernel Coefficient for 'rbf', 'poly', and 'sigmoid')





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# Final Model

For time purposes we used the Logistic Regression, with 'C': [10], 'penalty': ['I2'], 'solver': ['liblinear'], as the SVM Hyperparameter Tuning is a long process when started it would not have time or computing power to work with it.

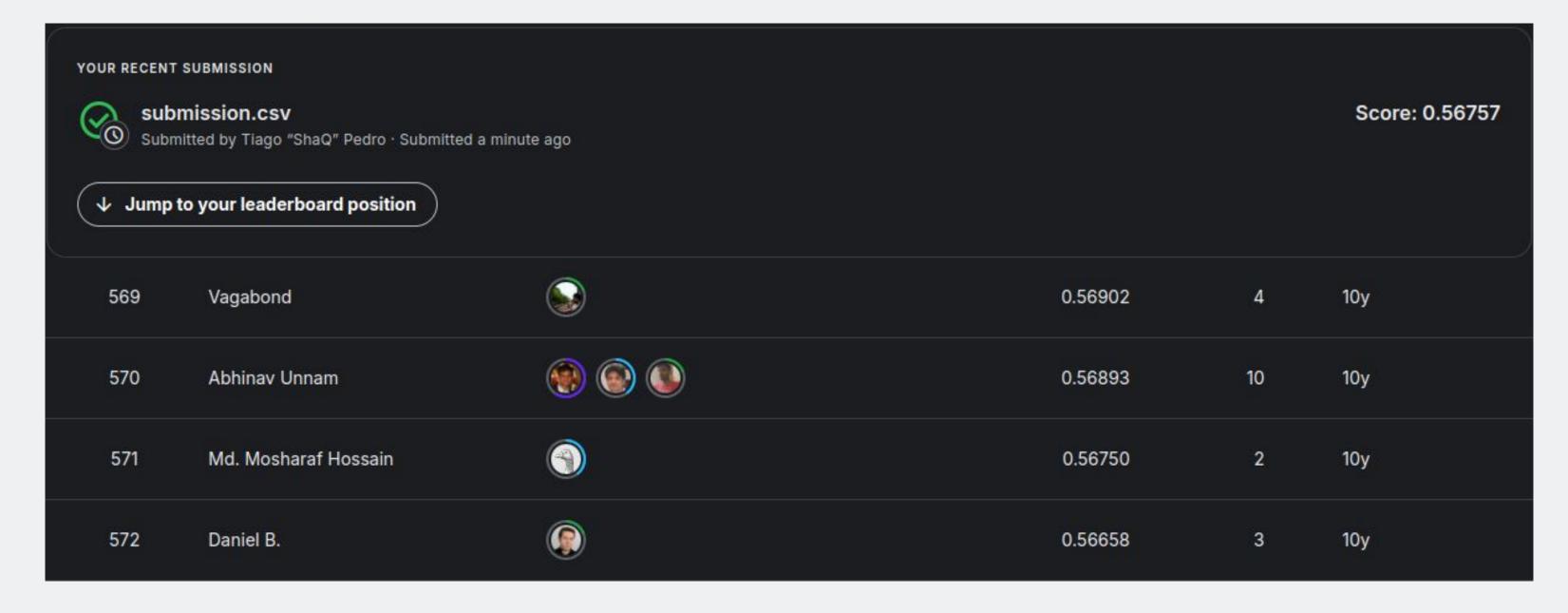
```
param_grid_lr = {
'C': [10],
'penalty': ['12'],
'solver': ['liblinear']
```







# Kaggle Submission









# Key Insights & Challenges



### **Machine Learning Models**

Choosing the right ones for the project and the ones possible to use



### **Tokenization**

The way we can compile a text to different type of values to input on the ML model



### Computing power struggle

Some ML models take longer to train or are even impossible to use with the current computing power



## **Hyperparameter Tuning**

Time consuming to try various ML, the need to test just the right from the 'standart' tests





# Possible Improvements

- Use of Transformers-Based models like BERT/DistilBERT
- Use of Deep Learning Models like RNNs or CNNs
- More computing power to be able to use the models with the full data set
- Test with different Tokenizers like KERAS
- Test different word embeddings like GloVe or Word2Vec





# Conclusion

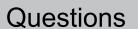
Machine learning is not only about the model we chose, but the way we manage the data before training the models, in this case using text to predict a data, there are various ways to treat them with different libraries for Tokenization, the lower score on the competition show the need to study and understand these libraries to optimize the data for better results.



Reporters

Introduction

Conclusion









# Thank you for your time and attention,ask any questions you may have

**Project made by Tiago Pedro**